

TOPICS COVERED

(NB: This list is not intended to be a complete list...)

Solving linear systems

↳ Gauss-Jordan Elimination

↳ Row reduction and RREF matrices

Geometry and linear systems

↳ dot product / angle formula.

Matrices and Matrix operations

↳ addition, scalar multiplication, matrix product, transpose

Vector spaces

↳ subspaces and subspace test

↳ span and linear independence

↳ * Bases and dimension

$S \subseteq V$ n/
 $S \neq \emptyset$ and $ax+by \in S$
for all scalar a, b and
all $x, y \in S$. $\rightarrow S \leq V$

$S \subseteq V$ is lin. ind. when
 $\sum c_i s_i = 0 \Rightarrow c_i = 0$ for all i

Linear maps

↳ linearity condition

↳ kernel and range spaces

(≈ null and column spaces)

↳ injectivity and surjectivity.

↳ Matrix representation

↳ Rank-Nullity Theorem $\rightarrow \text{rank}(L) + \text{nullity}(L) = \dim(\text{dom}(L))$

↳ Linear operators * $L: V \rightarrow V$

L is inj
iff $\ker(L) = \{0\}$

More on Matrices

↳ determinant

↳ elementary matrices *

↳ inversion of matrices

* Change of Basis

Eigenspaces

- ↳ Characteristic polynomial
- ↳ eigenvalues and eigenvectors
- ↳ Complex vector spaces

Diagonalization of matrices / linear operators

- ↳ Similar matrices $\rightsquigarrow B = PAP^{-1}$
- ↳ diagonalizability. $\rightsquigarrow M = \underset{\uparrow}{P} D P^{-1}$

Orthogonality (in \mathbb{R}^n).

- ↳ orthogonal projection
- *↳ orthogonal complement

$$\left[\text{Col}(M)^\perp = \text{null}(M^T) \right]$$

- ↳ Gram-Schmidt process *

- ↳ orthogonal matrices

$$\underline{A^{-1} = A^T} \quad (\text{i.e. } A^T A = I)$$

Symmetric Matrices $\rightsquigarrow A^T = A$

- *↳ Transpose $\rightsquigarrow (AB)^T = B^T A^T, (A+B)^T = A^T + B^T \dots$

- *↳ Real symmetric matrices have all eigenvalues real.

- ↳ Orthogonal diagonalizability

$$M = Q D Q^T$$

for Q orthogonal, D diagonal.

$$\rightarrow \boxed{\begin{array}{l} M \text{ symmetric iff} \\ M \text{ ortho. diag'able.} \end{array}}$$

$$\ker(L) = \text{kernel of a linear map.}$$

$$= \{v \in \text{dom}(L) : L(v) = 0\}$$

$$\text{null}(M) = \text{null space of matrix } M$$

$$= \text{solution set to } M\vec{x} = 0$$

$$= \ker(L_M) \quad \text{where} \quad \text{Rep}_{\mathcal{E}_n, \mathcal{E}_m}(L_M) = M.$$

\equiv

Point: kernel is associated to a linear map, whereas null space is associated to a matrix.

↳ often to compute a kernel of a linear map, we first compute the null space of an associated matrix, and then we convert that back into a kernel.

Ex: The linear map $L: \mathcal{P}_3(\mathbb{R}) \rightarrow \mathbb{R}^3$ given by

$$L(a_0 + a_1x + a_2x^2 + a_3x^3) = \begin{pmatrix} a_0 + a_1 \\ a_1 + a_2 \\ a_2 + a_3 \end{pmatrix}.$$

to compute $\ker(L)$, we will compute null space of an associated matrix. Let $B = \{1, x, x^2, x^3\} \subseteq \mathcal{P}_3(\mathbb{R})$.

w.r.t. B , L is represented by:

$$\left[[L(1)]_{\mathcal{E}_3} \mid [L(x)]_{\mathcal{E}_3} \mid [L(x^2)]_{\mathcal{E}_3} \mid [L(x^3)]_{\mathcal{E}_3} \right]$$

$$= \begin{bmatrix} \overset{a_0}{1} & \overset{a_1}{1} & \overset{a_2}{0} & \overset{a_3}{0} \\ 0 & 1 & 0 & 1 \\ 1 & 0 & 0 & 1 \end{bmatrix} = M.$$

$$\text{null}(M) = \text{null} \begin{bmatrix} 1 & 1 & 0 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 0 & 0 & 1 \end{bmatrix} = \text{null} \begin{bmatrix} 1 & 1 & 0 & 0 \\ 0 & 1 & 0 & 1 \\ 0 & -1 & 0 & 1 \end{bmatrix} = \text{null} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

$$= \text{null} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \rightsquigarrow \begin{cases} a_0 = 0 \\ a_1 = 0 \\ a_2 = t \\ a_3 = 0 \end{cases} \leftarrow$$

$$\therefore \ker(L) = \{v \in \mathcal{P}_3(\mathbb{R}) : a_0 + a_1x + a_2x^2 + a_3x^3 = v\}$$

$$a_0 = 0, a_1 = 0, a_2 = t \in \mathbb{R}, a_3 = 0$$

indicates

$$= \{tx^2 : t \in \mathbb{R}\} \subseteq \mathcal{P}_3(\mathbb{R}). \leftarrow$$

by contrast: $\text{null} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} = \left\{ \begin{bmatrix} 0 \\ 0 \\ t \\ 0 \end{bmatrix} \in \mathbb{R}^4 : t \in \mathbb{R} \right\} \subseteq \mathbb{R}^4$

M an $m \times n$ matrix can represent a linear map
from an n -dimensional space to an m -dimensional space.

Ex: Let $M = \begin{bmatrix} 2 & 2 & 4 \\ 1 & 0 & 1 \\ 3 & 0 & 2 \end{bmatrix}$.

\hookrightarrow Rank: $\text{null}(M) = \ker(L_M)$ where $L_M: \mathbb{R}^n \rightarrow \mathbb{R}^m$
has $L_M(x) = Mx$.

and $\text{col}(M) = \text{ran}(L_M)$ (same L_M as above)

Sol: To compute those, we compute $\text{RREF}(M)$ because:

$\text{null}(M) = \text{null}(\text{RREF}(M))$ AND

$\text{col}(M)$ has basis the columns of M corresp to
leading 1's in $\text{RREF}(M)$.

$$\begin{array}{ccccccc} \downarrow & \downarrow & \downarrow & & & & \downarrow & \downarrow & \downarrow \\ \begin{bmatrix} 2 & 2 & 4 \\ 1 & 0 & 1 \\ 3 & 0 & 2 \end{bmatrix} & \rightsquigarrow & \begin{bmatrix} 1 & 0 & 1 \\ 2 & 2 & 4 \\ 3 & 0 & 2 \end{bmatrix} & \rightsquigarrow & \begin{bmatrix} 1 & 0 & 1 \\ 0 & 2 & 2 \\ 0 & 0 & -1 \end{bmatrix} & \rightsquigarrow & \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{bmatrix} & \rightsquigarrow & \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \end{array}$$

$\therefore \text{null}(M) = \text{null}(I_3)$ has $\begin{cases} x = 0 \\ y = 0 \\ z = 0 \end{cases}$ so $\text{null}(M) = \left\{ \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \right\}$

has basis $\underline{\underline{\emptyset}}$. $\text{col}(M)$ has basis $\left\{ \begin{bmatrix} 2 \\ 1 \\ 3 \end{bmatrix}, \begin{bmatrix} 3 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 4 \\ 1 \\ 2 \end{bmatrix} \right\}$.

→ Can't be simplified... row operations change column spaces...

Ex: $\begin{bmatrix} 1 & 0 & 1 \\ -1 & 5 & -2 \\ 0 & 5 & -1 \end{bmatrix} \leftarrow$

Remark:

$$\text{nullity}(L_M) = 0 = \dim(\text{null}(M))$$

$$\text{rank}(L_M) = 3 = \dim(\text{col}(M))$$

$$\text{so } \dim(\mathbb{R}^3) = 3 = 0 + 3 = \text{nullity}(L_M) + \text{rank}(L_M)$$

→ $\text{nullity}(L_M) = 1$, $\text{rank}(L_M) = 2$. (check...).

L is injective when for all $x, y \in \text{dom}(L)$

we have $L(x) = L(y)$ implies $x = y$. *

→ "distinct inputs map to distinct outputs"

→ $L: \underline{\underline{V}} \rightarrow \underline{\underline{W}}$ is injective if and only if $\ker(L) = 0$.
linear.

L is surjective when for all $y \in \text{col}(L)$ there is an $x \in \text{dom}(L)$ such that $L(x) = y$.

→ "every element of the codomain is an output".

→ Rank-Nullity Thm: $\text{rank}(L) + \text{nullity}(L) = \dim(\text{dom}(L))$.

if $\text{rank}(L) = \dim(\text{col}(L))$, then L is surjective.

L is bijective when it is both surjective and injective.

→ Linear L is bijective iff L is an isomorphism.